



The 2nd Visual Object Tracking Segmentation VOTS2024 Challenge Results

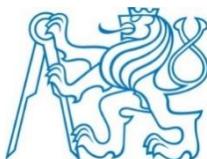
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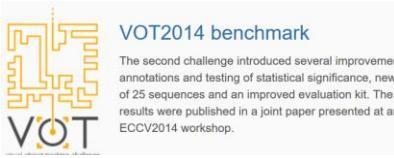
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VOTS2024 challenge

- The VOT formed in 2013 to support **general object trackers**
- Tracking objects that **do not change their topology**
(don't break apart or turn inside-out)



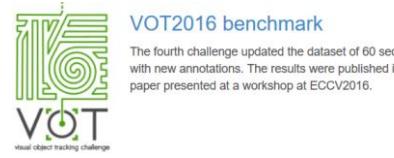
VOT2013 benchmark
The first challenge introduced a new evaluation kit plus 16 well-known short videos, 27 single-target trackers submitted by 51 participants participated at the challenge. The results were published in a joint paper presented at an ICCV2013 workshop which was attended by over 70 researchers.



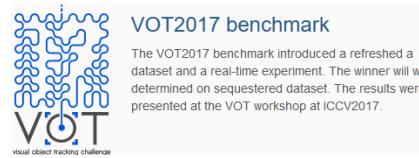
VOT2014 benchmark
The second challenge introduced several improvements in annotations and testing of statistical significance, new set of 25 sequences and an improved evaluation kit. The results were published in a joint paper presented at an ECCV2014 workshop.



VOT2015 benchmark
The third challenge introduced a dataset of 60 challenging sequences, a formalized sequence selection methodology and improvements to evaluation methodology. The results were published in a joint paper presented at an ICCV2015 workshop.



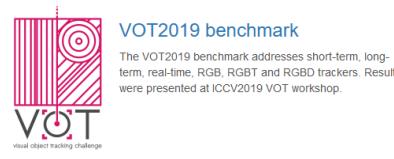
VOT2016 benchmark
The fourth challenge updated the dataset of 60 sequences with new annotations. The results were published in a joint paper presented at a workshop at ECCV2016.



VOT2017 benchmark
The VOT2017 benchmark introduced a refreshed a dataset and a real-time experiment. The winner will be determined on sequestered dataset. The results were presented at the VOT workshop at ICCV2017.



VOT2018 benchmark
The VOT2018 benchmark introduced a long-term subchallenge VOT-LT2018. Results were presented at the VOT workshop at ECCV2018.



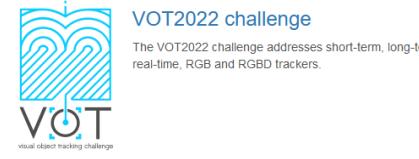
VOT2019 benchmark
The VOT2019 benchmark addresses short-term, long-term, real-time, RGB, RGBT and RGBD trackers. Results were presented at ICCV2019 VOT workshop.



VOT2020 benchmark
The VOT2020 benchmark addresses short-term, long-term, real-time, RGB, RGBT and RGBD trackers. Results were presented at the ECCV2020 VOT workshop.



VOT2021 challenge
The VOT2021 challenge addresses short-term, long-term, real-time, RGB and RGBD trackers. Results will be presented at the ICCV2021 VOT workshop.



VOT2022 challenge
The VOT2022 challenge addresses short-term, long-term, real-time, RGB and RGBD trackers.

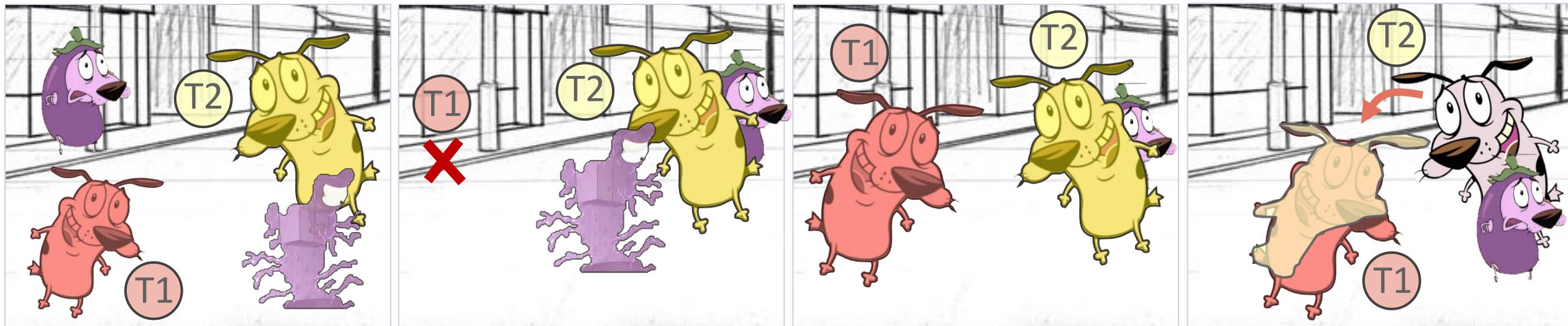
- In 2023 VOT challenges converged into **VOTS**:
segmentation-based general object tracking no longer
distinguishing between single-target, multi-target,
short-term and long-term tracking



The VOTS2024 challenge scope



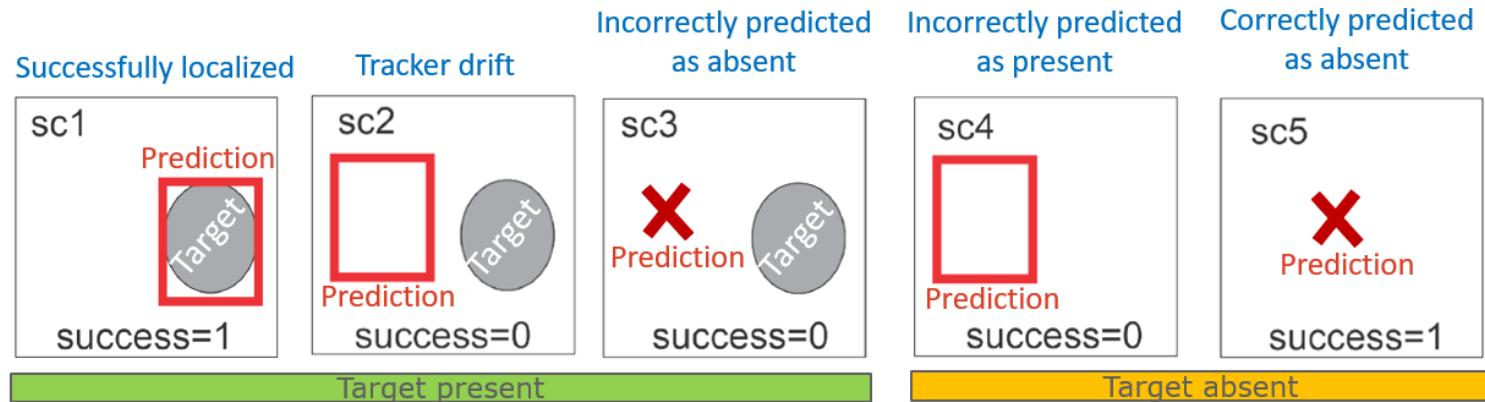
- General object Short/Long-term, Single/Multi-target segmentation trackers
- Initialize on all targets in the first frame and report position in the rest



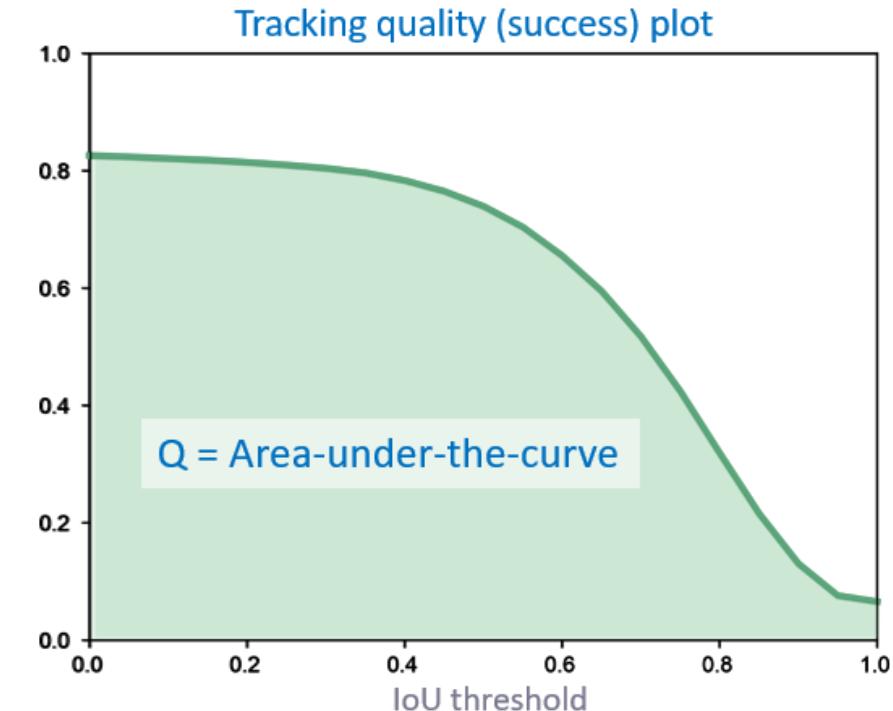
- Determine the **target absence** and **redetect** when it reappears
- **Drifting** off the target to **background or another object** is considered failure

VOTS Primary performance measure Q

- VOTS2023 *tracking success* accounts for 5 scenarios:

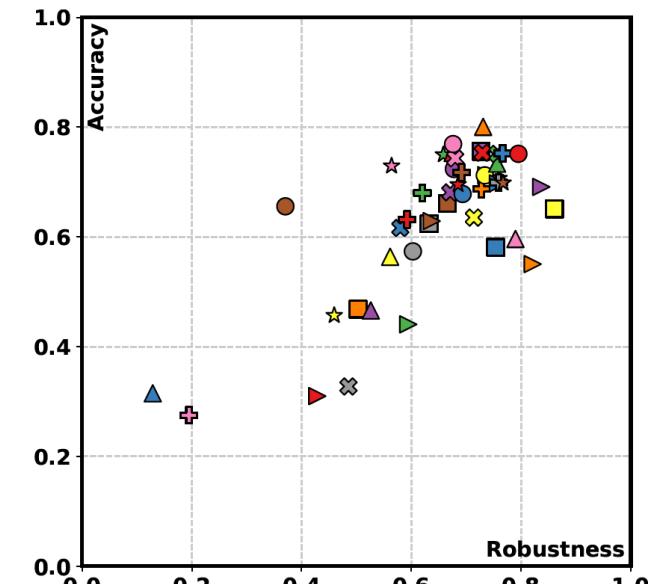
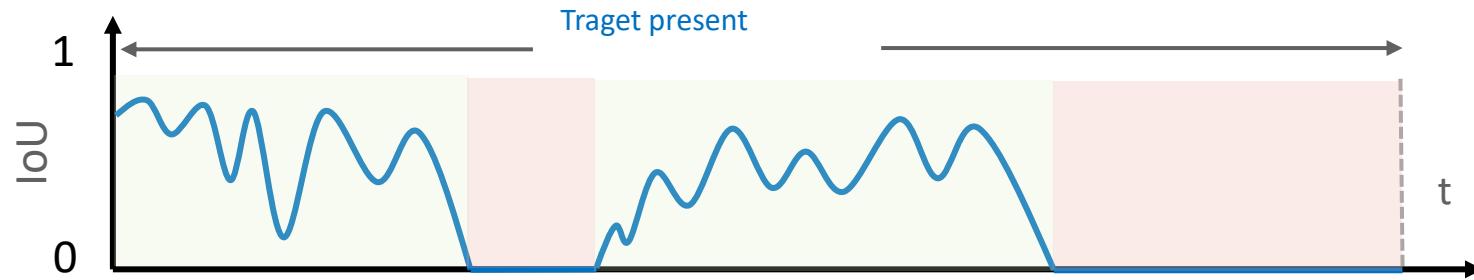


- Tracking quality plot:
 - Average of success plots calculated for each target in *each sequence*
- Primary measure: *Tracking quality Q* (area-under-the-curve)



VOTS Auxiliary performance measures

- Accuracy/R robustness (@IoU=0.0 when target present)



“Why did the tracker fail while target visible?”

- Not Reported Error (NRE): % frames incorrectly predicted target absent
- Drift Rate Error (DRE): % frames tracker drifted while predicting target present

“How well is target absence determined?”

- Absence Detection Quality (ADQ): % frames target correctly predicted absent

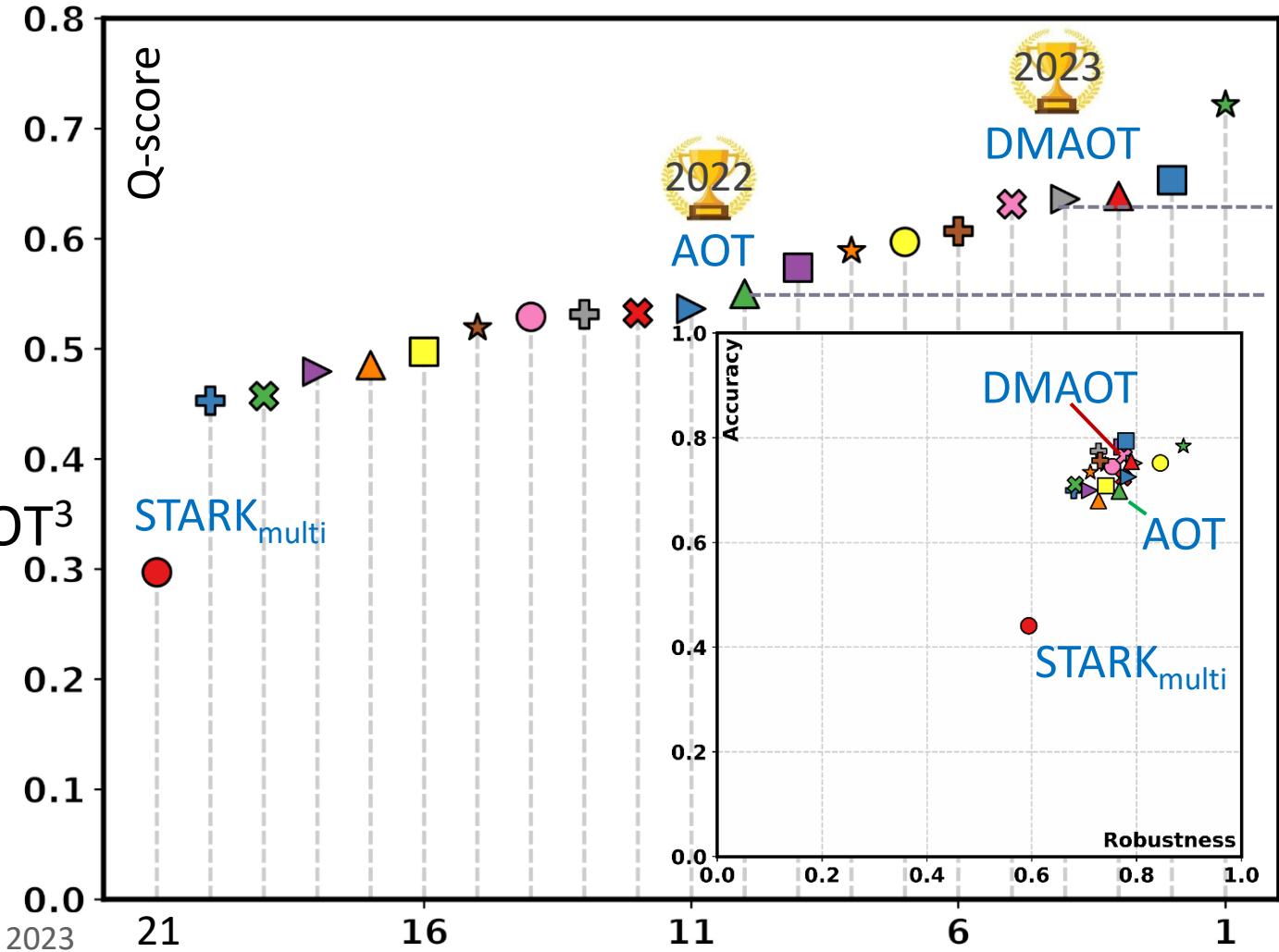
VOTS2024 dataset: same as VOTS2023

- Stats: 144 sequences ; 341 targets ; 168 targets leave the FOV at least once
- Sequence properties:
 - min/max = 63/10.7k frames
 - On average 2.37 targets per sequence annotated
 - Median target absence: 18 frames
- To prevent overfitting:
 - Sequences + initialization frames GT publicly available.
 - GT of test frames sequestered, evaluation carried out on a dedicated server.



VOTS2024 challenge results: 18 trackers tested

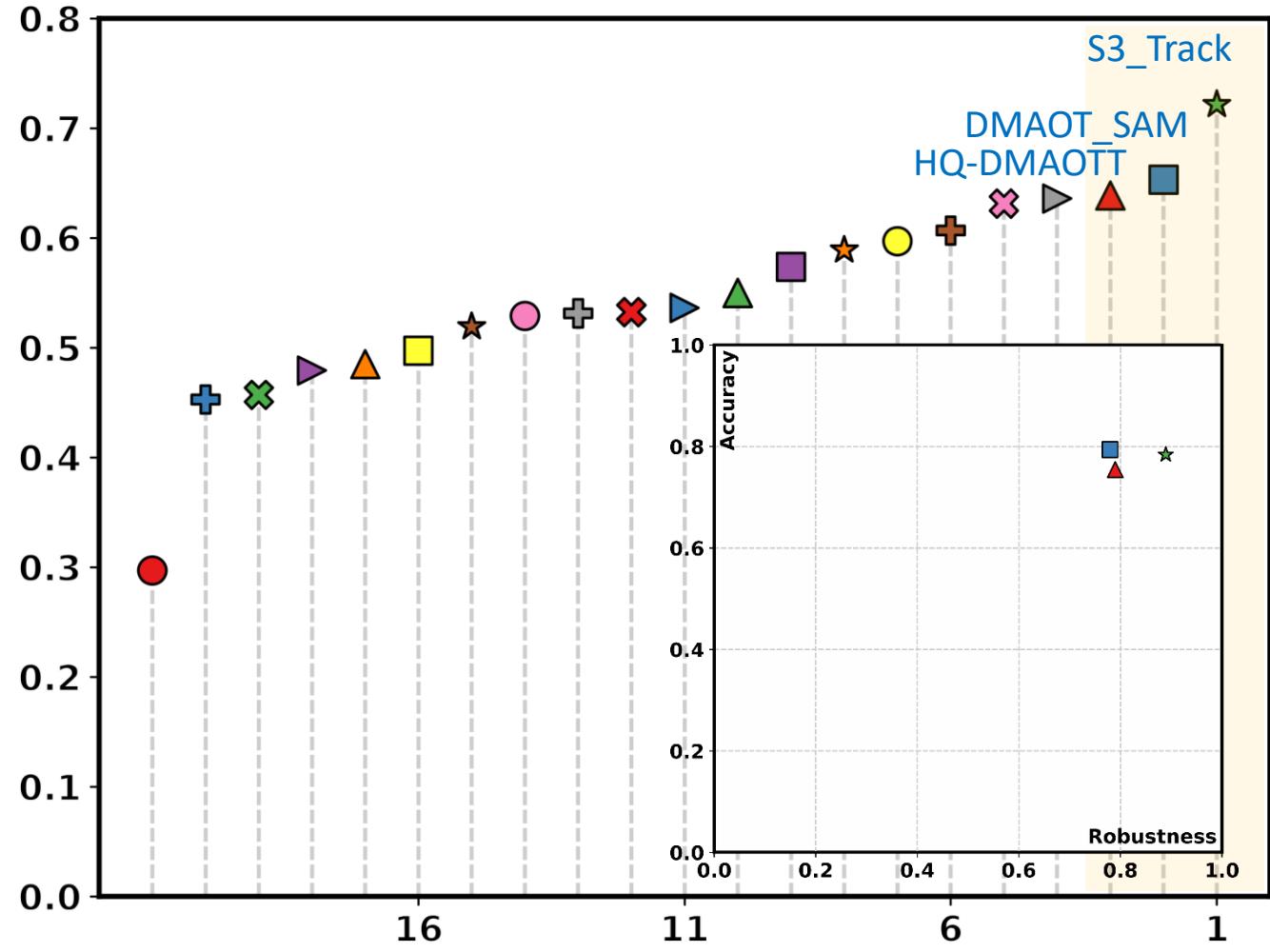
- Baseline 1: Independent STARKs¹ (41% in Q-score w.r.t. top tracker)
all submissions outperform it
- Baseline 2: VOT2022 winner AOT²
44% (8 trackers) outperfrom it
- Baseline 3: VOTS2023 winner DMAOT³
17% (3 trackers) outperfrom it



¹ Yan, et al. ICCV2021; ² Yang, et al. NeurIPS 2021 ; ³ Cheng, et al. VOTS 2023

VOTS2024 challenge top 3 trackers

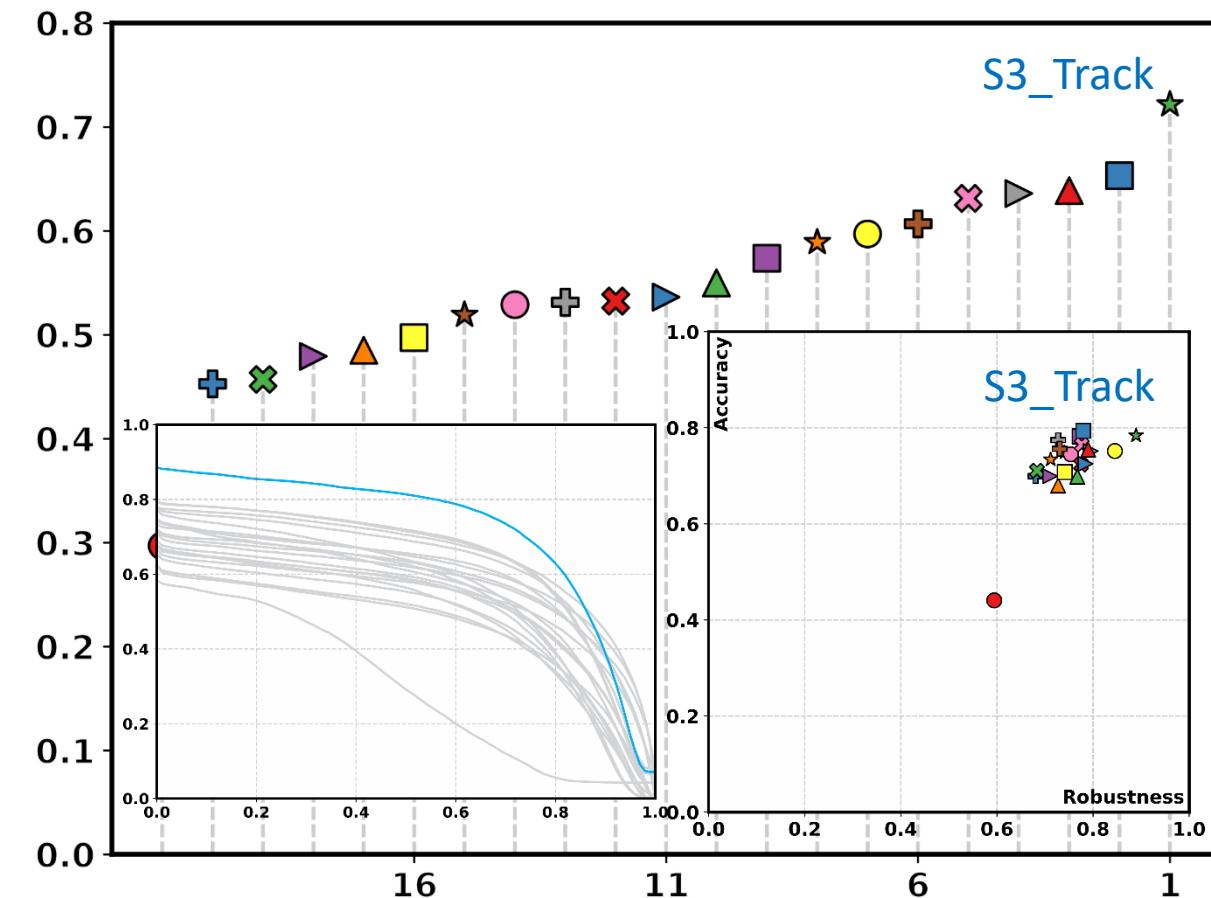
- Top 3 trackers: **S3_Track**, **DMAOT_SAM**, **HQ-DMAOTT**
 - **S3_Track:**
 - Single-stage architecture
 - Extends Cutie¹ by multi-scale features + ViT backbone +...
 - **DMAOT_SAM & HQ-DMAOTT:**
 - Two-stage architecture
 - DMAOT¹ for the initial mask prediction, then refine by SAM³
- All apply: **ViT/Swin** backbones + large training datasets



¹ Cheng, et al. CVPR2024; ² Cheng, et al. VOTS 2023; Kirillov et. al. ICCV2023

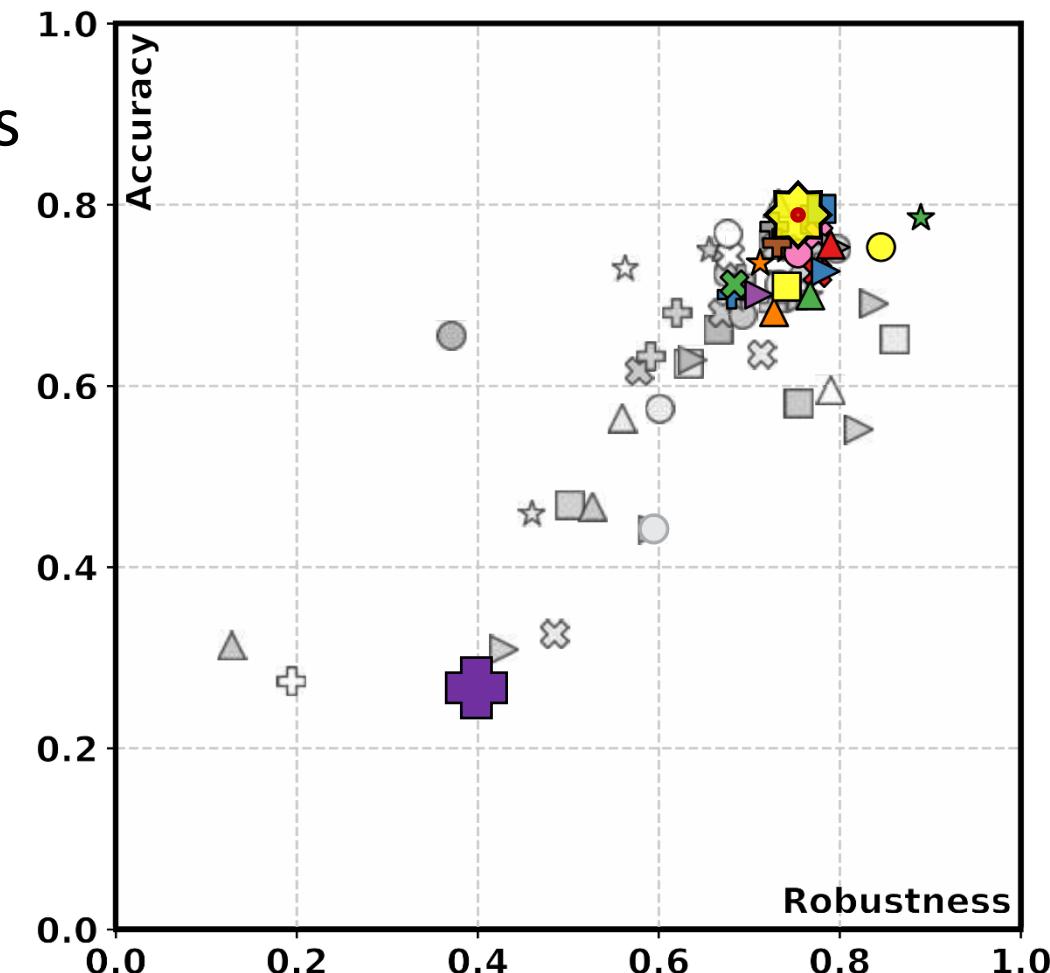
VOTS2024 top-ranked tracker S3_Track

- *Q-score* = 0.722 (>10% improvement over second-best)
- VOTS aux measures: $A=0.784$, $R=0.889$, $DRE=0.04$, $NRE=0.07$, $ADQ=0.781$
- Good segmentation accuracy 78.4% (*DMAOT_SAM*; $A=79.4\%$)
- On-target in ~89% of frames
- Localization failure 11% due to:
 - Falsely reported target absent: 63%
 - Drifting: 37%
- Accurately predicts target absence in 78% cases



VOTS2024 vs VOTS2023

- Compared to 2023, 2024 submissions improved on accuracy as well as robustness
 - VOTS2023 winner Q-score: 0.636
 - VOTS2024 top Q-score: 0.722
(13% winner performance boost)
- “Ancient” history (10 years):
- 2014 sota KCF¹: Q=0.107, A=0.268, R=0.403
- SAM2²: Q=0.642, A=0.789, R=0.75
- Outperforms last-year’s winner by 1% in Q
- Would have ranked 3rd in Q-score on VOTS2024



¹Henriques, et al. ECCV2012; ²Ravi, et al. Arxiv2024



VOTS2024 challenge

Winners:

S3_Track by: Deshui Miao, Xingsen Huang, Xin Li, Dongmei Jiang,
Ming-Hsuan Yang, Yaowei Wang

“Associating Semantic Tracker”

Winners talk end of Session I
@10:15

Summary

- Presented VOTS2024 results
- Evaluation server open for post-challenge evaluation



VOTS2024 benchmark

Thanks

- The VOTS2024 committee



M. Kristan



J. Matas



P. Tokmakov



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T. Tran



Xuan-Son Vu



Johanna Björklund



H. J. Chang



G. Fernandez

- Everyone who participated or contributed

Minasadat Attari, Antoni Chan, Liang Chen, Xin Chen, Jaired Collins, Yutao Cui, Ganesh Sai Manas Devarapu, Yinglong Du, Heng Fan, Wan-Cyuan Fan, Zhenhua Feng, Mingqi Gao, Rama Krishna Sai Gorthi, Raghav Goyal, Jungong Han, Bijaya Hatuwal, Zhenyu He, Xiantao Hu, Xingsen Huang, Yuqing Huang, Dongmei Jiang, Ben Kang, Palaniappan Kannappan, Josef Kittler, Simiao Lai, Ning Li, Xiaohai Li, Xin Li, Cheng Liang, Liting Lin, Haibin Ling, Ting Liu, Ziyan Liu, Huchuan Lu, Yifei Luo, Deshui Miao, Juan Mogollon, Ziqi Pang, Jaswanth Reddy Pochimireddy, Viktor Prutyanov, Gani Rahmon, Aleksandr Romanov, Liangtao Shi, Mennatullah Siam, Leonid Sigal, Arun Kumar Sivapuram, Roman Solovyev, Elham Soltani Kazemi, Imad Eddine Toubal, Jia Wan, Limin Wang, Xinying Wang, Yaowei Wang, Yu-Xiong Wang, Zhiquan Wang, Gangshan Wu, Qiangqiang Wu, Xiaojun Wu, Zihao Xia, Jinxiu Xie, Chenlong Xu, Tianyang Xu, Yong Xu, Chaocan Xue, Chao Yang, Jinyu Yang, Ming-Hsuan Yang, Chenyang Yu, Ke Yu, Chunhui Zhang, Jiaming Zhang, Zhipeng Zhang, Feng Zheng, Yaozong Zheng, Bineng Zhong, Jinglin Zhou, Junbao Zhou, Yong Zhou, Zikun Zhou, Guibo Zhu, Jiawen Zhu, Xuefeng Zhu, and Vladimir Zunin

- VOTS2024 sponsors:



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